

# Numerical Computation of Newton's Method

We shall discuss how to solve Nonlinear Systems of Equations using Newton's method. We will further demonstrate how to compute the Jacobian using automatic differentiation. Convergence results of Newton's method if any exist for our problem shall be discussed. All these shall be done in Julia.jl.

- NB: Reference is herein made to the Scientific Computing Lecture by Dr. Jurgen Fuhrmann (WS 21/22) for some coding idea.

## Basic Idea

This method is basically based on the idea of linear approximation. The tangent line at a given point is a good approximation of the function. Newton's iteration approximates a function based on its tangent line, then takes the tangent line's zero as the next guess  $\phi_i$ .

We begin with an initial (suitable) guess  $\phi_0$ . Then,  $\phi_0$  is used in to produce  $\phi_1$  and  $\phi_1$  is further used to produced  $\phi_2$  and so on. The objective is that one will ultimately find a value  $\phi_i$  that is close enough to the desired solution.

Recall the Taylor Series approximation of polynomials of increasing powers;

$$f(\phi) = f(\phi_0) + (\phi - \phi_0)f'(\phi_0) + \frac{1}{2}(\phi - \phi_0)^2f''(\phi_0) + \dots$$

where  $f'$  and  $f''$  are the first and second derivative of  $f$  at  $\phi_0$ . If we consider the first two terms of the Taylor series expansion we have:

$$f(\phi) = f(\phi_0) + f'(\phi_0)(\phi - \phi_0)$$

setting  $f(\phi) = 0$ ,

$$0 = f(\phi_0) + (\phi - \phi_0)f'(\phi_0)$$

$$(\phi - \phi_0)f'(\phi_0) = -f(\phi_0)$$

$$(\phi - \phi_0) = \frac{-f(\phi_0)}{f'(\phi_0)}$$

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In general, Newton's iteration in 1D is given by

$$\phi_{i+1} = \phi_i - \frac{f(\phi_i)}{f'(\phi_i)}$$

## • In multidimension:

We extend Newton's iteration in 1D into multi-dimensional system.

Essentially, we are finding the zeroes of  $F(\phi) = b$

where,  $F(\phi) := A\phi + \Psi(\phi)$  and  $b = 0$ .

In this case,  $F(\phi)$  is regarded as a nonlinear operator  $F : D \rightarrow \mathbb{R}^n$  where  $D \subset \mathbb{R}^n$  is the domain of definition,  $\phi = [\phi_1, \phi_2, \dots, \phi_n]^T \in \mathbb{R}^n$  and  $b = [0, 0, \dots, 0]^T \in \mathbb{R}^n$  is a zero vector.

The scalar  $\phi_i$  in the 1D case is replaced by vector  $\phi_i$  and we left multiply  $F(\phi_i)$  by the inverse of it's Jacobian matrix  $F'(\phi_i) = J(\phi_i)$ .

This results in a multidimensional Newton's iteration equation:

$$\phi_{i+1} = \phi_i - J(\phi_i)^{-1}F(\phi_i)$$

The inverse of  $F(\phi)$  is expensive to compute. Hence instead of computing  $J(\phi)^{-1}$ , we solve a linear equation by rearranging the iteration equation

$$J(\phi_i)\delta_i = -F(\phi_i)$$

where  $\delta_i = \phi_{i+1} - \phi_i$  using an initial guess  $\phi_0 \in \mathbb{R}^n$  and  $\delta_i \in \mathbb{R}^n$ .

we can then find  $\delta_i$  such that,

$$\delta_i = -J(\phi_i)^{-1}F(\phi_i)$$

Hence we replace  $-J(\phi_i)^{-1}F(\phi_i)$  with  $\delta_i$  in the iteration equation such that

$$\phi_{i+1} = \phi_i + \delta_i$$

## Step one : Inital Vector

The first step of Newton's method is to determine an initial vector. A standard way of determining an initial vector will be to compute the intersection of all three functions. Let  $\phi_0$  be our initial vector,

$$\phi_0 = [0.1 \quad 0.1 \quad -0.1]^T$$

and let  $\phi^*$  be the solution to the system  $F(\phi) = b$ .

```
phi_0 = ▶ [0.1, 0.1, -0.1]
```

```
• phi_0 = [0.1,0.1,-0.1]
```

## Step Two : Define the System

Next, we define the components of the nonlinear system  $F(\phi) = b$ . We define first the left hand vector  $F(\phi)$  followed by the right hand vector  $b$ , which is a zero vector.

```
F (generic function with 1 method)
```

```
• F(phi) = G(phi)
```

```
G (generic function with 1 method)
```

```
• function G(phi)
•     [3*phi[1] - cos(phi[2]*phi[3]) - 1/2;
•     phi[1]^2 - 81*(phi[2] + 0.1)^2 + sin(phi[3]) + 1.06;
•     exp(-phi[1]* phi[2]) + 20*phi[3] + (10*pi - 3)/3]
• end
```

```
b = ▶ [0, 0, 0]
```

```
• b = [0,0,0]
```

## Step Three : Evalaute $F$ at $\phi_0$

Now, we evaluate  $F(\phi)$  at the inital vector  $\phi$ .

```
▶ [-1.19995, -2.26983, 8.46203]
```

```
• F(phi_0)
```

After evaluating  $F(\phi_0)$ , we obtained a row vector. By transposing the row vector, we have the following column vector:

$$F(\phi_0) = \begin{bmatrix} -1.19995 \\ -2.26983 \\ 8.46203 \end{bmatrix}$$

## Step four : Residual

Here, the goal is to obtain the residual of the system. We compute the residual at the initial vector  $\phi_0$ .

```
► [-1.19995, -2.26983, 8.46203]

• begin
•   r(ϕi) = F(ϕi) - b    #i = 0,...n
•   r(ϕ0)
• end
```

We transpose the compute the residual above and obtain the following column vector

$$r(\phi_0) = \begin{bmatrix} -1.19995 \\ -2.26983 \\ 8.46203 \end{bmatrix}$$

## Step five : Jacobian

We compute and evaluate the Jacobian of  $F(\phi)$  at  $\phi_0$ . We denote the Jacobian at  $\phi_0$  by  $J(\phi_0)$ . This is done using Julia's automatic differentiation package "ForwardDiff: jacobian".

To begin with, we create a result buffer for our  $3 \times 3$  nonlinear system in order to use automatic differentiation to compute the Jacobian. We use the DiffResults.JacobianResult package. This is saved as d\_results.

Next, we call the ForwardDiff.jacobian! package on  $F$  with the initial vector  $\phi_0$  as input data. This mutates the  $d_{results}$ . Additionally, within the  $d_{results}$ , we obtain the value for the differential results. Finally, we obtain the Jacobian at  $\phi_0$ , which is a  $3 \times 3$  matrix as well as it's inverse  $J^{-1}(\phi_0)$ .

```
► [-1.19995, -2.07017, 12.462]

• begin
•   d_result=DiffResults.JacobianResult(ϕ0)
•   ForwardDiff.jacobian!(d_result,F,[0.1, 0.1, 0.1])
•   DiffResults.value(d_result)
• end
```

```
3×3 Matrix{Float64}:
 3.0      0.000999983  0.000999983
 0.2     -32.4        0.995004
-0.099005 -0.099005    20.0
```

```
• begin
•   J(ϕ0) = DiffResults.jacobian(d_result)
•   J(ϕ0)
• end
```

```
3×3 Matrix{Float64}:
 0.333332  1.03404e-5  -1.71808e-5
 0.0021086 -0.0308688      0.00153563
 0.00166051 -0.000152757  0.0500075
```

```
• inv(J(ϕ0))
```

The results are written explicitly below, where the Jacobian is given by:

$$J(\phi_0) = \begin{bmatrix} 3.0 & 0.000999983 & 0.000999983 \\ 0.2 & -32.4 & 0.995004 \\ -0.099005 & -0.099005 & 20.0 \end{bmatrix}.$$

and the inverse of the Jacobian is given as,

$$J(\phi_0)^{-1} = \begin{bmatrix} 0.333332 & 1.03404e-5 & -1.71808e-5 \\ 0.0021086 & -0.0308688 & 0.00153563 \\ 0.00166051 & -0.000152757 & 0.0500075 \end{bmatrix}.$$

## Solving the System

All the above steps are brought together in the Newton function below. The function below is called on  $F, b$  and the initial vector  $\phi_0$  at a tolerance level of  $1.0e-12$  for an iteration limit of 100.

Newton (generic function with 1 method)

```

• function Newton(F,b,ϕ₀; tol=1.0e-12, maxit=100)
•     result=DiffResults.JacobianResult(ϕ₀)           #result buffer for Jacobian
•     history=Float64[]                               #history vector
•     ϕ=copy(ϕ₀)
•     it=1
•     while it<maxit
•         ForwardDiff.jacobian!(result,(ϕ)->F(ϕ) - b,ϕ)
•         res=DiffResults.value(result)               #residual result
•         jac=DiffResults.jacobian(result)             #Jacobian result
•         δ=jac\res                                    #solve for δ
•         ϕ= ϕ - δ                                     #update
•         nm=norm(δ)                                   #norm
•         push!(history,nm)                            #push norm to history
•         if nm<tol
•             return ϕ,history
•         end
•
•         it=it+1
•     end
•     throw("convergence failed")
• end

```

For our outcomes, we have the following functions:

- the approximate solution obtained after the iteration is given by the function `Newton_result`
- `Newton_history` shows results for convergence
- And residual for the residual results.

```

▼ (
1: ▶ [0.5, 8.10963e-18, -0.523599]
2: ▶ [0.586567, 0.0179945, 0.00157676, 1.24488e-5, 7.76083e-10, 8.90637e-17]
3: ▶ [0.0, -2.22045e-16, 1.77636e-15]
)
• begin
•     Newton_result,Newton_history=Newton(F,b,ϕ₀)
•     residual = F(Newton_result)-b
•     Newton_result,Newton_history,residual
• end

```

## Convergence Result

The measure of convergence is checked by computing the norm  $\|\phi_{i+1} - \phi_i\|$  in every iteration step and the tolerance level.

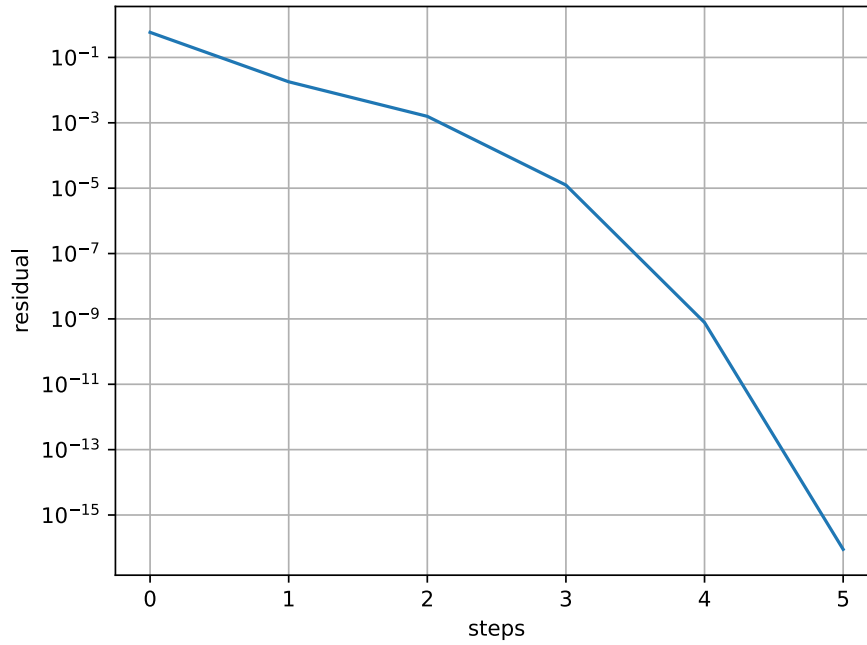
Once the first approximate solution vector  $\phi_1$  is computed, the iteration process continues until we obtain convergence to the solution  $\phi^*$ .

For convergence of a set of vectors, the norm must be zero. That is,

$$\|\phi_{i+1} - \phi_i\| = \sqrt{(\phi_{i+1}^1 - \phi_i^1)^2 + \dots + (\phi_{i+1}^n - \phi_i^n)^2} = 0.$$

Therefore, Newton's iteration function will terminate when the norm is zero and less than the level of tolerance.

The convergence of our iteration is visualized below. We observe from the below that, in fewer iteration steps, we have a good behaviour of the residual. That is, in **6** steps, we obtain a quadratic convergence which is desirable.



## Discussion

Iteration steps: 6

Using the Newton function, we obtained the result for our system  $F(\phi) = b$  in six iteration steps. The iteration stopped due to the fact that, we obtained convergence as explained earlier.

Thus, the approximate solution of the system from our computation is written explicitly as:

$$\phi^* = \begin{bmatrix} 0.5 \\ 8.10963e - 18 \\ -0.523599 \end{bmatrix}$$